

SMBUD Project - MongoDB

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1 Introduction

In this chapter will be presented the problem specification and the hypothesis under which the database is implemented.

1.1. Problem Specification

This project aims to build a documental database that handles scientific articles contained in the DBLP bibliography. The focus is on creating a database which allows efficient information retrieval of the articles, including their chapters and images. The main collections analyzed in the project are *Authors* and *Publication* with all their attributes and related objects like: *chapters*, *biographies* and *images*.

1.2. Assumptions

- 1. Articles can be published on a single venue
- 2. There is no distinction between different types of Publication
- 3. An author can't work for more than one organization for the same Publication
- 4. A chapter can contain subsections and these will be considered chapters as well



2 | ER Diagram

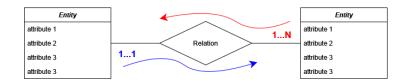


Figure 2.1: ER Diagram Organization

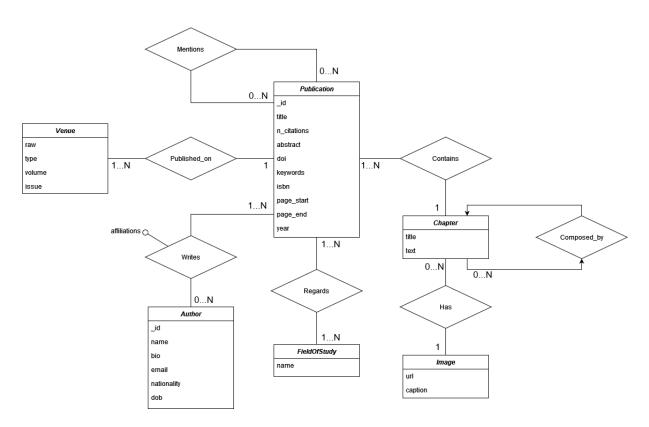


Figure 2.2: ER Diagram

4 2 ER Diagram

The Entity-Relationship model contains 6 main entities, that are related to each other through various relationships:

- **Publication:** represents all the scientific articles. Its attributes are: _id, title, n_citations, abstract, doi, keywords, isbn, page_start, page_end, year and its organization will be presented later
- Author: it is the one who contributed to a publication. Its attributes are: _id, name, bio, email, nationality, dob (date of birth)
- Venue: it is where a publication is published or presented. Its attributes are: raw, type, volume, issue
- **FieldOfStudy:** this entity represents the subjects of the publication and its attribute is *name*
- Chapter: it represents the chapter of a scientific articles. Its attributes are title, text
- Image: it represents the image contained in a specific chapter of the publication.

 Its attributes are url, caption

The ER diagram designed contains also the following relationships:

- Mentions: is the relationship between a *publication* and another *publication* cited by the first one
- Published on: is the relationship between a publication and its venue
- Writes: is the relationship between *author* and *publication* which features the affiliation property. We decided to design it with affiliation as an attribute of the relationship, due to the fact that it belongs only to a pair of *author* and *publication* and it represents the institute where the author worked for the publication
- Regards: is the relationship between a publication and its fields of study
- Contains: is the relationship between publication and its chapters
- Composed_by: is the relationship between two *chapters*. It represents the relationship created between a chapter and its sections, between a section and its subsection and so on. Note that we used a directed arrow in order to indicate that a section belongs only to a chapter, but a chapter could own more than one section.
- Has: is the relationship between *chapter* and its *images*

3 Document Structure

In this section we will describe the structure of our database. We decided to split our dataset in two collections: *authors* and *articles*. The reason behind this choice was to increase the performance and to reduce the spatial complexity. For example if we need to modify the email of an author we just have to change the field in the *author* collection. Whereas, if we just kept one collection, very expensive update query would have been needed.

Furthermore, to avoid redundancies we used manual references to bind the two collections.

In order to show the number of Authors and Articles in our collections, we performed the two following queries:

3.1. Data Preprocessing

Starting from the database sample used for the Neo4j part some operations were performed to add the required fields and to exploit better mongoDB functionalities.

Missing isbn, page_start, page_end attributes were added with random generations.

Most important, chapters were added from the original paper pdf when available, and parsing was performed by using PHP Library Smalot PDFParser. This library only parses the full text of the papers so the text was then divided into chapters using chapter titles as a reference to split, thus it was possible to obtain the 2 attributes' title and text

for each chapter (for the text field, only 10 rows were taken into consideration). A total of 170 pdfs were parsed and that was also used as a pool to pick random chapters for papers that could not be parsed because their pdfs were not available as a free copy on aminers' site.

To add images and captions, a dataset of films and their covers from tmdb was used, for every chapter a random number of images links were added combined with movie titles used as a caption.

The advantage of aminers' dataset was the built-in _id attribute for articles and authors and also the existence of an array of references with the list of the referenced _id inside every article tuple.

To better exploit this opportunity we believe it was handy to have the 2 collections of articles and authors, so we performed JSON parsing of the dump to generate 2 new JSON files:

- one that contains the articles, with an array of authors that has inside the references to their _id objectId
- the other contains the authors with an array of articles that has inside the references to their _id objectId.

In the authors' collection the bio attribute was randomly generated only when not available, while email, nationality and dob attributes were 100% randomly generated, especially:

- email: was generated by concatenating Name.Surname with a pool of most famous mail servers(gmail.com, yahoo.com...)
- nationality: was generated by picking randomly from a pool of 20 nation code
- dob date of birth: was generated randomly from 1940 to 1992, so it may be possible that some inconsistencies are present (e.g. author that published before having 20 years old or even before being born)

3.2. Article structure

In the following JSON file we can see how the Article is structured in our dataset.

Note: in some fields we decided not to insert all the content only for ease of read reasons.

Note: in venue, fields issue, volume and publisher are missing because they are empty in this example.

```
1 {
2 "_id": {
    "$oid": "53e99f86b7602d9702859fdf"
4 },
5 "title": "Locality Sensitive Outlier Detection: A ranking driven
     approach",
6 "authors": [
    "id": {
      "$oid": "542a4c9fdabfae61d496694e"
9
10
    "org": "Computer Science and Engineering Department, OhioSU, USA"
    },
12
    {
13
    "id": {
      "$oid": "53f48bc5dabfaea7cd1cce1d"
15
16
    "org": "Computer Science and Engineering Department, OhioSU, USA"
17
    },
18
19
    "id": {
2.0
      "$oid": "53f44b6fdabfaec09f1dd00d"
21
22
    "org": "Computer Science and Engineering Department, OhioSU, USA"
23
    }
24
25],
"n_citation": 60,
27 "abstract": "Outlier detection is fundamental to a variety of database
     and . . . " ,
"doi": "10.1109/ICDE.2011.5767852",
29 "keywords": [
    "database point",
    "ranking scheme",
31
    "geometric approach",
33
    . . .
34],
35 "isbn": "978-1-4244-8958-9",
```

```
36 "page_start": "410",
37 "page_end": "421",
38 "year": 2011,
39 "fos": [
    "Locality-sensitive hashing",
    "Anomaly detection",
41
    "Machine learning",
42
43
44],
45 "venue": {
    "raw": "ICDE",
   "type": 0,
48 },
49 "chapters": [
    "title": "1. Introduction",
51
    "text": "Open Computing Language (OpenCL)[3] is a unified programming
52
     . . . " ,
    "images": [
54
      "caption": "Giraffada",
55
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
      {
58
      "caption": "A Certain Magical Index: The Miracle Of Endymion",
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
61
62
      "caption": "Ready To Rumble",
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
65
      "caption": "Raw Force",
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
68
      },
69
      {
70
      "caption": "The Departed",
71
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
72
      },
73
      "caption": "Repeaters",
75
      "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
76
      }
78
```

```
},
79
    "title": "2. The OpenCL Framework",
81
    "text": "21 Organization of Our Runtime\nThe target cluster...",
82
    "images": [
84
       "caption": "The King Of New York",
85
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
88
       "caption": "The Ghost Who Walks",
89
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
91
92
       "caption": "Fighter In The Wind",
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
94
       },
95
       {
96
       "caption": "Wake Up",
97
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
98
      },
99
       "caption": "Infierno Blanco",
101
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
102
      },
       "caption": "Bajo El Mismo Techo",
105
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
107
       {
108
       "caption": "Le Dernier Samaritain",
109
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
110
      }
    ]
112
    },
113
114
    "title": "3. Evaluation",
115
    "text": "We have implemented the OpenCL runtime and...",
    "images": [
118
       "caption": "L'hotel Degli Amori Smarriti",
119
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
      },
```

```
"caption": "Bit",
123
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
124
       },
126
       "caption": "Dead Silence",
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
128
       },
129
130
       "caption": "Amenazados",
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
132
133
    ]
    },
135
136
     "title": "4. Conclusions",
     "text": "We introduce the design and implementation of...",
138
     "images": [
139
       {
       "caption": "Eddie The Eagle",
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
142
       },
143
144
       "caption": "Gorenos",
145
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
146
       },
147
148
       "caption": "Pinocchio",
149
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
       },
       {
152
       "caption": "Le Grand Bazar",
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
       },
156
       "caption": "Gangsters",
157
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
       },
       "caption": "Kodachrome",
161
       "link": "https://image.tmdb.org/t/p/w600_and_h900_bestv2/....jpg"
163
164
     }
166],
```

3.3. Author structure

In the following JSON file we can see how the Author is structured in our dataset.

Note: in some fields we decided not to insert all the content only for ease of read reasons.

3.4. Attributes Description

In this section we will present all the attributes contained in our filtered dataset.

3.4.1. Article

Publication represent the central concept of the system and contains:

- id is an ObjectId that identifies a publication.
- title represents the title of the publication.
- authors is an array of subdocuments that contains: ObjectId of the authors of the article and the org field which represent the affiliation.
- n citation is the number of times that the publication has been mentioned.
- abstract is a string containing a brief summary of the contents of the paper.
- doi Digital Object Identifier is a persistent and standardized identifier.
- keywords is an array containing keywords of the publication.
- **isbn** is an identification code of the venue of the publication.
- page start defines the starting page of the publication.
- page end defines the last page of the publication.
- year represents the year of publication.
- fos is an array containing the fields of study of the publication.
- **venue** is a sub-document that represents where a publication is published or presented. This field contains:
 - raw is the name or the abbreviation of the venue (regardless the year, issue or volume) in which the publication was presented.
 - type indicates the type of the publication.
 - **volume** is the volume of the venue in which the article has been published.
 - issue refers to how many times a periodical has been published during that year.
 - **publisher** is the name of the publisher of the article.

- **chapters** is an array of sub-documents containing containing **title** (title of the chapter), **text** (the content), **sections** (that has the same structure of a chapter). They also contains **images** composed by:
 - caption the caption of the image.
 - link the link to the image.
- references set of ObjectIds of the referenced articles.

3.4.2. Author

The dataset provides the following author fields:

- **id** is an ObjectId that identifies an author.
- name is the name of the author.
- articles is a set of articles identifier of the publications of the author.
- **orcid** Open Researcher and Contributor ID is a unique identifier for authors of scientific articles.
- bio is a string that describes the author.
- email is the email address of the author.
- **nationality** is the nationality of the author.
- **dob** is the birth date of the author.



4 Commands and Queries

4.1. Commands

We have identified the following INSERT and UPDATE commands to show the system basic functionalities.

4.1.1. Insert a publication in the system

Assuming it is not present in the dataset, we inserted a new document that is a new instance of *Publication*. In order to do that we instantiated 5 different variables: article_id that generates an ObjectId representing the id of the article we're creating; article_ref1 and article_ref2 that are the ids of two scientific articles cited by this publication; author1_id and author2_id that are the ids of the authors.

Note: isbn field is missing

Note: type = 1 represents a Journal

Note: issue field is missing

```
article_id = ObjectId()
2 article_ref1 = ObjectId("53e99fe4b7602d97028bf743")
article_ref2 = ObjectId("53e99fddb7602d97028bc085")
4 author1_id = ObjectId()
5 author2_id = ObjectId()
7 db.articles.insertOne({
    _id: article_id,
    title: "An extensive study of C-SMOTE, a Continuous Synthetic Minority
      Oversampling Technique for Evolving Data Streams",
    authors:
      {id:author1_id, org:"Politecnico di Milano"},
12
      {id:author2_id, org:"Politecnico di Milano"}
13
14
    n_citation: 3,
```

```
abstract: "Streaming Machine Learning (SML) studies algorithms that
     update their models, given an unbounded and often non-stationary flow
      of data performing a single pass. Online class imbalance learning is
      a branch of SML that combines the challenges of both class imbalance
      and concept drift. In this paper, we investigate the binary
     classification problem by rebalancing an imbalanced stream of data in
      the presence of concept drift, accessing one sample at a time.",
    doi: "10.1016/j.eswa.2022.116630",
17
    keywords: ["Evolving Data Stream", "Streaming", "Concept drift", "
18
     Balancing"],
    page_start: 39,
19
    page_end: 46,
    year: 2022,
2.1
    fos: ["Computer Science", "Stream Reasoning", "Big Data"],
22
      raw: "ESA",
24
      type: 1,
25
      volume: 196,
      publisher: "Elsevier"
    },
28
    chapters:
29
30
      title: "1. Introduction",
32
      text: "Nowadays, data abound as a multitude of smart devices, such as
33
      smartphones, wearables, computers, and Internet of Things (IoT)
     sensors produce massive, continuous, and unbounded flows of data,
     namely data streams. This poses several challenges to Machine
     Learning (ML)."
      },
34
35
36
      title: "2. Background",
      text: "This section is divided into three parts describing the
38
     different concept drift types and characteristics, the evaluation
     metrics, and the most common approaches to use in class imbalance.",
      subsection: [
40
        title: "2.1. Concept drift in evolving data streams",
41
        text: "In this part, we introduce the concept drift phenomenon
     explaining why and how it happens. We explain all its different types
     , forms, and possible speeds of occurrence.",
        subsubsection: [
43
44
```

```
title: "2.1.1. Concept drift types",
45
          text: "Since the generating function is unknown, concept drift is
      unpredictable. In the batch settings, with all the data available,
     it is simple to check and detect if a dataset is not stationary.",
          images:
             Γ
48
               {
49
               caption: "Fig.1 Representation of the three different types
     of concept drift",
               url: "https://ars.els-cdn.com/content/image/1-s2.0-
51
     S0957417422001208-gr1.jpg"
            ٦
53
          }
54
        ]
        }
56
57
      },
      title: "3. C-SMOTE",
61
      text: "This section recalls the description of C-SMOTE, inspired by
62
     the Smote technique, originally presented in Bernardo, Gomes et al.
     (2020). C-SMOTE is designed to rebalance an imbalanced data stream,
     and it can be pipelined with any SML- model. C-SMOTE stands for
     Continuous-Smote, meaning that the new Smote version is applied
     continuously.",
      images:[
63
        {
        caption: "Fig. 2. Architecture of C-SMOTE meta-strategy pipelined
65
     with an Online Learner.",
        url: "https://ars.els-cdn.com/content/image/1-s2.0-
66
     S0957417422001208-gr5.jpg"
        }
67
      ],
68
      subsection:[
69
        title: "3.1. Artificial data streams",
7.1
        text: "To synthetically reproduce the different types of concept
72
     drifts shown in Section 2.1, we choose two of the most commonly used
     artificial data generators: SINE1 (Gama et al., 2004) and SEA (Street
      & Kim, 2001)."
        }
74
```

```
75
76 }
77 ],
78 references: [
79 article_ref1,
80 article_ref2
81 ]
82
83 })
```

4.1.2. Insert an author in the system

Assuming he is not present in the dataset, we used insertOne to create a new instance of Author.

Note: author1_id and article_id, refers to the variables instantiated in the previous command (Section: 4.1.1)

```
db.authors.insertOne({
    _id: author1_id,
    name: "Emanuele Della Valle",
    orcid: "0000-0002-5176-5885",
    articles:[
5
      article_id
   ],
    bio: "Emanuele Della Valle holds a PhD in Computer Science from the
     Vrije Universiteit Amsterdam and a Master degree in Computer Science
     and Engineering from Politecnico di Milano. He is associate professor
      at the Department of Electronics, Information and Bioengineering of
     the Politecnico di Milano.",
    email: "emanuele.dellavalle@gmail.com",
   nationality: "it",
10
    dob: ISODate ("1975-03-07T00:00:00.000Z")
12 })
```

4.1.3. Update the number of citations of referenced publications

With the following snippet of code is possible to increment the $n_{citations}$ field of the *Publications* referenced by the article created in section 4.1.1.

Note: in this command we used updateMany in order to update both the referenced publications; update wasn't enough because only one of the two matching document would have been updated

4.1.4. Modification of the biography of an author

This command allows to access Authors by field name in order to append a string to field bio

4.1.5. Add a publication to its author

This command allows to add the new publication (Publication creation presented in section 4.1.1), to one of its Authors.

Note: we assumed newArticleId as the identifier of the new publication and that author1_id refers to the variable instantiated in section 4.1.1

4.2. Queries

We have identified the following queries in order to show the system's basic functionalities. For ease of read reasons, we decided to show results obtained using project operator, and, in some cases, we showed only some of the results.

4.2.1. Query 1

This query returns one publication written after 2013 whose FieldOfStudy(fos) contains 'Machine learning'.

```
db.articles.findOne({
    "$and": [{year: {$gte:2013}}, {"fos":"Machine learning"}]
}
```

```
< { id: ObjectId("53e99f86b7602d970285ab95"),</pre>
   title: 'Bootstrapping polarity classifiers with rule-based classification',
   year: 2013,
    [ 'Bag-of-words model',
      'Rule-based system',
      'Bootstrapping',
      'Computer science',
      'Artificial intelligence',
      'Classifier (UML)',
      'Margin classifier',
      'Classifier (linguistics)',
      'Linear classifier',
      'Machine learning',
      'Quadratic classifier' ],
    [ { title: '1. Offline Learning of Multiple Regression Model' },
      { title: '2. Analysis on Online Time Complexity' },
      { title: '3. Analysis on Offline Time Complexity' } ] }
```

Figure 4.1: Projection on title, fos, year, chapters.title

4.2.2. Query 2

This query returns all the Italian authors born after 1960.

```
{ _id: ObjectId("53f43107dabfaee02ac8f08b"),
    name: 'WOLFGANG GLÄNZEL',
    articles: [ ObjectId("53e99fc9b7602d97028a40cd") ],
    orcid: '0000-0001-7529-5198',
    bio: 'Hideo Murakami ($\'73-M\'74) received the B.S. degree in electrical engineering from Kanazawa
    email: 'WOLFGANG.GLÄNZEL@yahoo.com',
    nationality: 'it',
    dob: 1968-10-23T00:00:00.0002 }

{    _id: ObjectId("53f42dbedabfaec09f1197b3"),
    name: 'Murari Mani',
    articles: [ ObjectId("53e99fd0b7602d97028abcd9") ],
    bio: 'Hiromi Narimatsu received the B.Sci. from Tsuda College, Japan, in 2009. She is now in master
    email: 'Murari.Mani@yandex.ru',
    nationality: 'it',
    dob: 1981-08-15T00:00:00.0002 }
```

4.2.3. Query 3

This query returns ten of the articles published by 'Elsevier' after 2009.

Description: this query perform a filtering over the year attribute of the article and on the publisher field of the venue sub-document.

Figure 4.2: Projection on title, fos, year, chapters.title, venue.raw, venue.publisher

4.2.4. Query 4

This query returns the top three years sorted by number of publications.

Description: the query computes the count by aggregating with respect to the year field. Then groups are sorted by descending order and only the top 3 are kept.

```
{ _id: 2013, count: 612 }
   { _id: 2010, count: 574 }
   { _id: 2011, count: 561 }
```

4.2.5. Query 5

This query finds all the articles with at least one Stanford affiliation and regarding 'Machine learning' field of study.

Description: this query perform a filtering over the fos attribute of the article and on the org field of the authors array of sub-documents.

Figure 4.3: Projection on authors, for and title

4.2.6. Query 6

This query returns the top three years sorted by number of distinct authors.

Description: the query starts by unwinding the authors array. Then, articles are grouped by year and by author. The distinct count over the year is performed grouping the previous results by year and accumulating 1 for each group. Finally, results are sorted by descending order, keeping only the top 3 years.

```
{ _id: 2013, count: 2016 }
   { _id: 2010, count: 1823 }
   { _id: 2011, count: 1750 }
```

4.2.7. Query 7

This query returns the 20 most frequent keywords.

Description: the query starts by unwinding the keywords array. The next stage groups with respect to the keywords (case insensitive) and computes the count for each group. Finally, the results are ordered by descending order and limited to the top 20

4.2.8. Query 8

This query, given the title of an article, returns the chapter with the highest number of images.

Description: the first stage matches the article given the title. Then an unwind of the chapters array is performed. The project stage is used to compute the variable **imgCount** which stores the number of images contained in each chapter. The results are sorted by descending order and limited to top 1.

4.2.9. Query 9

This query returns articles citing another article that contains, between its references, at least one article written in a Milan University.

Description: the first stage loads, by id, the array of cited documents into a new field

called refs. Then a match is used to find in this field all the articles that contain at least 1 author that worked for a Milan university.

Note: we used the operator lookup to issue a join, in order to access another article instance.

```
db.articles.aggregate([
    { $lookup: {
      from: "articles",
      localField: "references",
      foreignField: "_id",
      as: "refs"
      }
    },
    { $match: {
9
      "refs.authors.org":{$regex: "Milan"}
10
      }
    },
12
    {"$project": {
1.3
      "title":1,
14
      "refs.title":1,
      "refs.authors": 1
      }
    }
18
19 ])
```

Figure 4.4: Projection on title, refs.title, refs.authors

4.2.10. Query 10

This query finds the author with the maximum amount of written articles and retrieves his article with the highest number of coauthors.

Description: First of all, we compute the artCount field which contains the length of

the articles array field. Then we sort in descending order and keep only the first author by artCount. After that, a join is performed to load the articles documents into the new field articles_doc. A new projection is then used to compute the number of authors of each loaded article. Finally we sort the result in descending order and keep only the top 1 article. The projection is used to display the result in a clearer way.

Note: we used the operator lookup to issue a join in order to access articles of the author.

```
db.authors.aggregate([
    {"$project": {
      "_id":1,
      "name":1,
      "articles":1,
      "artCount": { "$size": "$articles" } } },
    {$sort : {"artCount": -1}},
    {$limit :1 },
    {$lookup: {
9
      from: "articles",
10
      localField: "articles",
      foreignField: "_id",
12
      as: "articles_doc"}},
13
    {"$unwind" : {path: "$articles_doc"}},
14
    {"$project": {
15
      "name":1,
      "articles_doc":1,
17
      "artCount":1,
18
      "authCount": { "$size": "$articles_doc.authors"}}},
19
    {$sort : { "authCount": -1}},
20
    {$limit : 1},
21
    {"$project": {
22
      "name":1,
23
      "articles_doc.title":1,
24
      "articles_doc.authors":1,
      "authCount": 1,
26
      "artCount":1 }}
28])
```

Figure 4.5: Projection on name, articles_doc.title, articles_doc.authors, authCount, art-Count

4.2.11. Query 11

This query returns all the articles written by authors whose names combined have all 26 letters of the alphabet.

Desciption: The query is done with the following steps:

- an unwind on articles is performed to obtain an entry for each article written by an author
- results are sorted by authors' name in order to have the list sorted alphabetically for further steps
- for each article, a group is performed and the authors list is pushed into an array
- a string concatenating all the authors names retrieved by article documents, is computed and converted to lowercase
- a map operation is performed to split letter by letter the entries; a filter with a regex to keep only alphabet characters is applied, then an unwind on letters is made
- entries are then grouped by article _id and authors list; letters are reduced into an array, then the size of this array is computed and only entries containing all 26 letters are matched
- finally, a join operation is performed on articles to obtain the title

```
db.authors.aggregate([
      "$unwind" : {path: "$articles"}
    },
    { "$sort" : { "name": 1 }},
      "$group" : {
        _id: "$articles",
        authors: {
           $push: {
1.0
             $concat: ["$name"]
11
          }
        }
13
      },
14
    },
16
17
      $project: {
        "_id": 1,
        "authors": "$authors",
        "fullNames": {
          $reduce: {
21
            input: "$authors",
22
            initialValue: "",
            in: { $toLower: {$concat : ["$$value", "$$this"]}}
24
          }
25
        }
      }
27
    },
2.8
29
      $project: {
30
         "_id": 1,
31
        "authors": "$authors",
32
        letters: {
           $filter: {
34
             input: {
35
               $map: {
                 input: {
                   $range: [ 0, { "$strLenCP": "$fullNames" } ]
3.8
                 },
                 in: {
                   "$substrCP":
41
42
                      "$fullNames",
                     "$$this",
```

```
1
                    ]
46
                  }
47
               }
             },
             cond: {
50
                $regexMatch: {
                  input: "$$this",
                  regex: '[a-z]'
53
             }
           }
        }
57
58
    },
    { $unwind: '$letters'},
60
61
       $group: {
62
         _id: {
63
           "_id": '$_id',
64
           authors: "$authors"
         },
         letters: { $addToSet: '$letters' },
      }
68
    },
69
      $project: {
71
         "_id" : 1,
72
         "authors": "$authors",
         "differentLetters": {
           "$size": "$letters"
         }
      }
78
    { $match : {"differentLetters" : 26}},
79
80
      $lookup: {
81
         from: "articles",
82
         localField: "_id._id",
83
         foreignField: "_id",
         as: "articles_doc"
85
      }
    },
    {
```

```
id: ObjectId("53e9a042b7602d9702929e98"),
title: [ 'Building an information retrieval test collection for spontaneous conversational speech' ],
 [ 'Bhuvana Ramabhadran',
   'Dagobert Soergel',
   'G. Craig Murray',
   'James Mayfield',
   'Jianqiang Wang',
   'Liliya Kharevych',
   'Samuel Gustman',
   'Stephanie Strassel',
title: [ 'Design principles for developing stream processing applications' ],
[ 'Bugra Gedik',
   'Deepak S. Turaga',
   'Henrique Andrade',
   'Jeffrey David Harris',
   'Olivier Verscheure',
   'Paul Jones',
   'William Szewczyk' ],
```

Figure 4.6: Projection on id, title, authors, differentLetters

4.2.12. Improving Queries Perfomance

In the queries, we performed string searches on relatively small textual fields using the \$\mathscr{s}regex\$ operator. This kind of search is easy to use and works very well on small datasets, but it is not optimal in large databases as it is not utilizing indexes efficiently. If we

wanted to perform advanced and high-performing full-text search queries, we would have had to define textual indexes among the two collections. Since only one textual index can be created over a collection, an index over multiple textual fields should be defined. These indexes can require some disk space and use a lot of resources when created.

5 Conclusions

The documental approach turned out to be very flexible and intuitive, thanks to its affinity with the object-oriented paradigm. Also, this kind of technology allowed us to shape data to match the most frequent operations that could take place in the publication domain. Therefore, queries became very simple and efficient.

Moreover, with our implementation, we tried to reach a good trade-off between performance and spatial complexity, avoiding choices that could lead to critical data duplication problems.